

Chapter 2

A Review on Differentiation

Reading: Spivak pp. 15-34, or Rudin 211-220

2.1 Differentiation

Recall from 18.01 that

Definition 2.1.1. A function $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is **differentiable** at $a \in \mathbb{R}^n$ if there exists a linear transformation $\lambda : \mathbb{R}^n \rightarrow \mathbb{R}^m$ such that

$$\lim_{h \rightarrow 0} \frac{|f(a+h) - f(a) - \lambda(h)|}{|h|} = 0 \quad (2.1)$$

The norm in Equation 2.1 is essential since $f(a+h) - f(a) - \lambda(h)$ is in \mathbb{R}^m and h is in \mathbb{R}^n .

Theorem 2.1.2. If $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is differentiable at $a \in \mathbb{R}^n$, then there is a **unique** linear transformation $\lambda : \mathbb{R}^n \rightarrow \mathbb{R}^m$ that satisfies Equation (2.1). We denote λ to be $Df(a)$ and call it the **derivative** of f at a

Proof. Let $\mu : \mathbb{R}^n \rightarrow \mathbb{R}^m$ such that

$$\lim_{h \rightarrow 0} \frac{|f(a+h) - f(a) - \mu(h)|}{|h|} = 0 \quad (2.2)$$

and $d(h) = f(a + h) - f(a)$, then

$$\lim_{h \rightarrow 0} \frac{|\lambda(h) - \mu(h)|}{|h|} = \lim_{h \rightarrow 0} \frac{|\lambda(h) - d(h) + d(h) - \mu(h)|}{|h|} \quad (2.3)$$

$$\leq \lim_{h \rightarrow 0} \frac{|\lambda(h) - d(h)|}{|h|} + \lim_{h \rightarrow 0} \frac{|d(h) - \mu(h)|}{|h|} \quad (2.4)$$

$$= 0. \quad (2.5)$$

Now let $h = tx$ where $t \in \mathbb{R}$ and $x \in \mathbb{R}^n$, then as $t \rightarrow 0$, $tx \rightarrow 0$. Thus, for $x \neq 0$, we have

$$\lim_{t \rightarrow 0} \frac{|\lambda(tx) - \mu(tx)|}{|tx|} = \frac{|\lambda(x) - \mu(x)|}{|x|} \quad (2.6)$$

$$= 0 \quad (2.7)$$

Thus $\mu(x) = \lambda(x)$. □

Although we proved in Theorem 2.1.2 that if $Df(a)$ exists, then it is unique. However, we still have not discovered a way to find it. All we can do at this moment is just by guessing, which will be illustrated in Example 1.

Example 1. Let $g : \mathbb{R}^2 \rightarrow \mathbb{R}$ be a function defined by

$$g(x, y) = \ln x \quad (2.8)$$

Proposition 2.1.3. $Dg(a, b) = \lambda$ where λ satisfies

$$\lambda(x, y) = \frac{1}{a} \cdot x \quad (2.9)$$

Proof.

$$\lim_{(h,k) \rightarrow 0} \frac{|g(a+h, b+k) - g(a, b) - \lambda(h, k)|}{|(h, k)|} = \lim_{(h,k) \rightarrow 0} \frac{|\ln(a+h) - \ln(a) - \frac{1}{a} \cdot h|}{|(h, k)|} \quad (2.10)$$

Since $\ln'(a) = \frac{1}{a}$, we have

$$\lim_{h \rightarrow 0} \frac{|\ln(a+h) - \ln(a) - \frac{1}{a} \cdot h|}{|h|} = 0 \quad (2.11)$$

Since $|(h, k)| \geq |h|$, we have

$$\lim_{(h,k) \rightarrow 0} \frac{|\ln(a+h) - \ln(a) - \frac{1}{a} \cdot h|}{|(h, k)|} = 0 \quad (2.12)$$

□

Definition 2.1.4. *The **Jacobian matrix** of f at a is the $m \times n$ matrix of $Df(a) : \mathbb{R}^n \rightarrow \mathbb{R}^m$ with respect to the usual bases of \mathbb{R}^n and \mathbb{R}^m , and denoted $f'(a)$.*

Example 2. *Let g be the same as in Example 1, then*

$$g'(a, b) = \left(\frac{1}{a}, 0\right) \quad (2.13)$$

Definition 2.1.5. *A function $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is **differentiable on** $A \subset \mathbb{R}^n$ if f is differentiable at a for all $a \in A$. On the other hand, if $f : A \rightarrow \mathbb{R}^m$, $A \subset \mathbb{R}^n$, then f is called **differentiable** if f can be extended to a differentiable function on some open set containing A .*

2.2 Properties of Derivatives

Theorem 2.2.1. 1. *If $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is a constant function, then $\forall a \in \mathbb{R}^n$,*

$$Df(a) = 0. \quad (2.14)$$

2. *If $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is a linear transformation, then $\forall a \in \mathbb{R}^n$*

$$Df(a) = f. \quad (2.15)$$

Proof. The proofs are left to the readers □

Theorem 2.2.2. *If $g : \mathbb{R}^2 \rightarrow \mathbb{R}$ is defined by $g(x, y) = xy$, then*

$$Dg(a, b)(x, y) = bx + ay \quad (2.16)$$

In other words, $g'(a, b) = (b, a)$

Proof. Substitute p and Dp into L.H.S. of Equation 2.1, we have

$$\lim_{(h,k) \rightarrow 0} \frac{|g(a+h, b+k) - g(a, b) - Dg(a, b)(h, k)|}{|(h, k)|} = \lim_{(h,k) \rightarrow 0} \frac{|hk|}{|(h, k)|} \quad (2.17)$$

$$\leq \lim_{(h,k) \rightarrow 0} \frac{\max(|h|^2, |k|^2)}{\sqrt{h^2 + k^2}} \quad (2.18)$$

$$\leq \sqrt{h^2 + k^2} \quad (2.19)$$

$$= 0 \quad (2.20)$$

□

Theorem 2.2.3. *If $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is differentiable at a , and $g : \mathbb{R}^m \rightarrow \mathbb{R}^p$ is differentiable at $f(a)$, then the composition $g \circ f : \mathbb{R}^n \rightarrow \mathbb{R}^p$ is differentiable at a , and*

$$D(g \circ f)(a) = Dg(f(a)) \circ Df(a) \quad (2.21)$$

Proof. Put $b = f(a)$, $\lambda = f'(a)$, $\mu = g'(b)$, and

$$u(h) = f(a+h) - f(a) - \lambda(h) \quad (2.22)$$

$$v(k) = g(b+k) - g(b) - \mu(k) \quad (2.23)$$

for all $h \in \mathbb{R}^n$ and $k \in \mathbb{R}^m$. Then we have

$$|u(h)| = \epsilon(h)|h| \quad (2.24)$$

$$|v(k)| = \eta(k)|k| \quad (2.25)$$

where

$$\lim_{h \rightarrow 0} \epsilon(h) = 0 \quad (2.26)$$

$$\lim_{k \rightarrow 0} \eta(k) = 0 \quad (2.27)$$

Given h , we can put k such that $k = f(a + h) - f(a)$. Then we have

$$|k| = |\lambda(h) + u(h)| \leq [\|\lambda\| + \epsilon(h)]|h| \quad (2.28)$$

Thus,

$$g \circ f(a + h) - g \circ f(a) - \mu(\lambda(h)) = g(b + k) - g(b) - \mu(\lambda(h)) \quad (2.29)$$

$$= \mu(k - \lambda(h)) + v(k) \quad (2.30)$$

$$= \mu(u(h)) + v(k) \quad (2.31)$$

Thus

$$\frac{|g \circ f(a + h) - g \circ f(a) - \mu(\lambda(h))|}{|h|} \leq \|\mu\|\epsilon(h) + [\|\lambda\| + \epsilon(h)]\eta(h) \quad (2.32)$$

which equals 0 according to Equation 2.26 and 2.27. \square

Exercise 1. (Spivak 2-8) Let $f : \mathbb{R} \rightarrow \mathbb{R}^2$. Prove that f is differentiable at $a \in \mathbb{R}$ if and only if f^1 and f^2 are, and that in this case

$$f'(a) = \begin{pmatrix} (f^1)'(a) \\ (f^2)'(a) \end{pmatrix} \quad (2.33)$$

Corollary 2.2.4. If $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$, then f is differentiable at $a \in \mathbb{R}^n$ if and

only if each f^i is, and

$$\lambda'(a) = \begin{pmatrix} (f^1)'(a) \\ (f^2)'(a) \\ \vdots \\ (f^m)'(a) \end{pmatrix}. \quad (2.34)$$

Thus, $f'(a)$ is the $m \times n$ matrix whose i th row is $(f^i)'(a)$

Corollary 2.2.5. *If $f, g : \mathbb{R}^n \rightarrow \mathbb{R}$ are differentiable at a , then*

$$D(f + g)(a) = Df(a) + Dg(a) \quad (2.35)$$

$$D(fg)(a) = g(a)Df(a) + f(a)Dg(a) \quad (2.36)$$

$$D(f/g)(a) = \frac{g(a)Df(a) - f(a)Dg(a)}{[g(a)]^2}, \quad g(a) \neq 0 \quad (2.37)$$

Proof. The proofs are left to the readers. □

2.3 Partial Derivatives

Definition 2.3.1. *If $f : \mathbb{R}^n \rightarrow \mathbb{R}$ and $a \in \mathbb{R}^n$, then the limit*

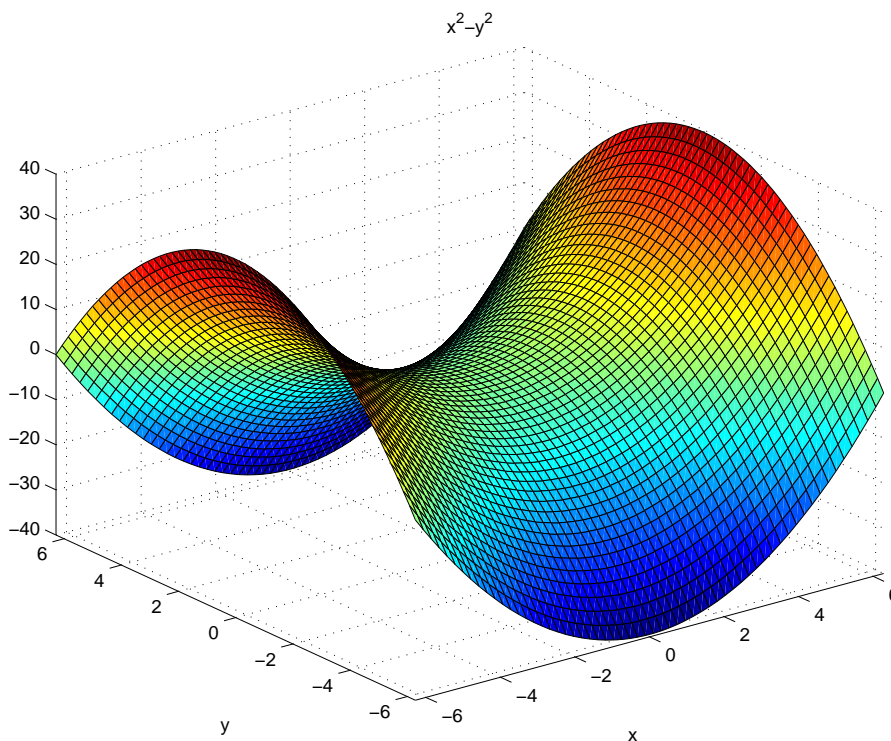
$$D_i f(a) = \lim_{h \rightarrow 0} \frac{f(a^1, \dots, a^i + h, \dots, a^n) - f(a^1, \dots, a^n)}{h} \quad (2.38)$$

*is called the i th **partial derivative** of f at a if the limit exists.*

If we denote $D_j(D_i f)(x)$ to be $D_{i,j}(x)$, then we have the following theorem which is stated without proof. (The proof can be found in Problem 3-28 of Spivak)

Theorem 2.3.2. *If $D_{i,j}f$ and $D_{j,i}f$ are continuous in an open set containing a , then*

$$D_{i,j}f(a) = D_{j,i}f(a) \quad (2.39)$$



Partial derivatives are useful in finding the extrema of functions.

Theorem 2.3.3. *Let $A \subset \mathbb{R}^n$. If the maximum (or minimum) of $f : A \rightarrow \mathbb{R}$ occurs at a point a in the interior of A and $D_i f(a)$ exists, then $D_i f(a) = 0$.*

Proof. The proof is left to the readers. \square

However the converse of Theorem 2.3.3 may not be true in all cases. (Consider $f(x, y) = x^2 - y^2$).

2.4 Derivatives

Theorem 2.4.1. *If $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is differentiable at a , then $D_j f^i(a)$ exists for $1 \leq i \leq m, 1 \leq j \leq n$ and $f'(a)$ is the $m \times n$ matrix $(D_j f^i(a))$.*